

Data Cleaning

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```
library(dplyr)
library(fpp)
library(fpp2)
library(seasonal)
library(ggplot2)

library(imputeTS) # imputation for ts
```

```
dailyweather = read.csv("DailyWeather.csv")
colnames(dailyweather)
```

```
## [1] "Year"      "Month"     "Day"       "sunshine"  "wind"      "rainfall"  "max"       "min"
```

Data exploration

```
summary(dailyweather)
```

```
##      Year      Month      Day      sunshine      wind      rainfall
## Min.   :1970   Min.    : 1.000   Min.    : 1.00   Min.    : 0.000   Min.    :  5.40   Min.    :  0.000
## 1st Qu.:1982   1st Qu.: 4.000   1st Qu.: 8.00   1st Qu.: 3.300   1st Qu.: 35.30   1st Qu.:  0.000
## Median :1995   Median : 7.000   Median :16.00   Median : 6.800   Median : 44.60   Median :  0.000
## Mean   :1995   Mean    : 6.529   Mean    :15.72   Mean    : 6.548   Mean    : 47.61   Mean    :  1.467
## 3rd Qu.:2007   3rd Qu.:10.000   3rd Qu.:23.00   3rd Qu.: 9.700   3rd Qu.: 59.40   3rd Qu.:  0.800
## Max.   :2020   Max.    :12.000   Max.    :31.00   Max.    :13.900   Max.    :139.00   Max.    :138.800
##                                     NA's    :10642   NA's     :64
```

More than 50% missing value in sunshine variable

Combine date columns

```
df_combinedate = dailyweather %>% mutate(date = as.Date(with(dailyweather, paste(Year,
↪   Month, Day, sep = "-")), "%Y-%m-%d")) %>% select(-c("Year",
  "Month", "Day"))
```

Missing values clustering index

```
col_sunshine = df_combinedate$sunshine
x = 0
for (i in 1:length(col_sunshine)) {
  if (is.na(col_sunshine)[i] == FALSE) {
    x = i
    break
  }
}

# sunshine value all equal to NA before 1999-08-18
df_combinedate[x, ]
```

```
##      sunshine wind rainfall  max min      date
## 10641      2.7 87.1        0 15.4 6.1 1999-08-18
```

```
df_remove_naclust = df_combinedate[-(1:x - 1), ] #df after 1999-08-18
summary(df_remove_naclust)
```

```
##      sunshine      wind      rainfall      max      min      date
## Min.   : 0.000   Min.   : 13.00   Min.   : 0.000   Min.   : 8.10   Min.   : -2.000   Min.   : 1999-0
## 1st Qu.: 3.300   1st Qu.: 35.30   1st Qu.: 0.000   1st Qu.:15.30   1st Qu.: 6.700   1st Qu.:2004-1
## Median : 6.800   Median : 44.60   Median : 0.000   Median :19.30   Median : 9.500   Median :2009-1
## Mean   : 6.548   Mean   : 47.32   Mean   : 1.394   Mean   :20.55   Mean   : 9.881   Mean   :2009-1
## 3rd Qu.: 9.700   3rd Qu.: 57.60   3rd Qu.: 0.600   3rd Qu.:24.40   3rd Qu.:12.800   3rd Qu.:2015-0
## Max.   :13.900   Max.   :122.40   Max.   :138.800   Max.   :46.80   Max.   :30.500   Max.   :2020-0
## NA's    :2       NA's    :25                                     NA's    :1
```

Remaining missing value index

```
## sunshine
col_sunshine1 = df_remove_naclust$sunshine
c = 1
y1 = NA
for (i in 1:length(col_sunshine1)) {
  if (is.na(col_sunshine1)[i] & c <= 2) {
    y1[c] = i
    c = c + 1
  }
}
```

```

}
y1 # 4067 7535

## wind
col_wind = df_remove_naclust$wind
c = 1
y2 = NA

for (i in 1:length(col_wind)) {
  if (is.na(col_wind)[i] & c <= 25) {
    y2[c] = i
    c = c + 1
  }
}
y2 #24 145 146 147 148 149 150 1908 2441 2442 2635 3109 3517 3616 3677 3798 4155
↪ 4354 4419 4425 4504 4520 5370 6374 6375

## min
y3 = NA
col_min = df_remove_naclust$min
for (i in 1:length(col_min)) {
  if (is.na(col_min)[i]) {
    y3 = i
    break
  }
}
y3 # 15

```

```

## [1] 4067 7535
## [1] 24 145 146 147 148 149 150 1908 2441 2442 2635 3109 3517 3616 3677 3798 4155 4354 4419 4
## [1] 15

```

Missing Value Imputation

```

ts_df = ts(df_remove_naclust, frequency = 365.25, start = c(1999, 8))

ts_imputed = na_interpolation(ts_df, option = "linear") # linear interpolation

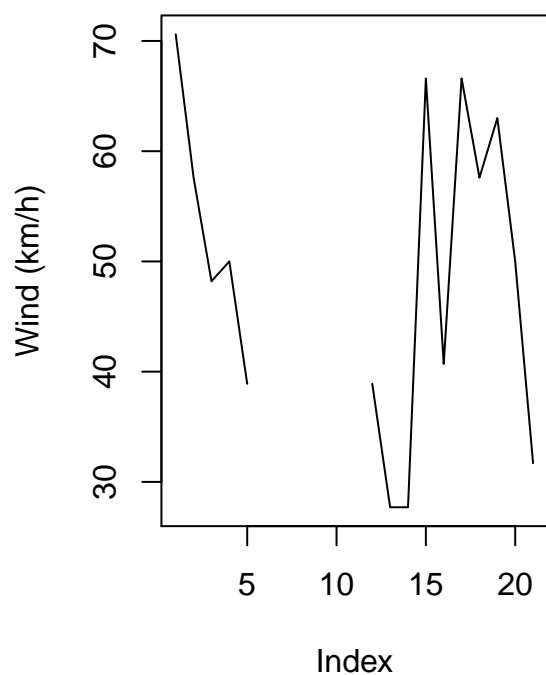
```

```

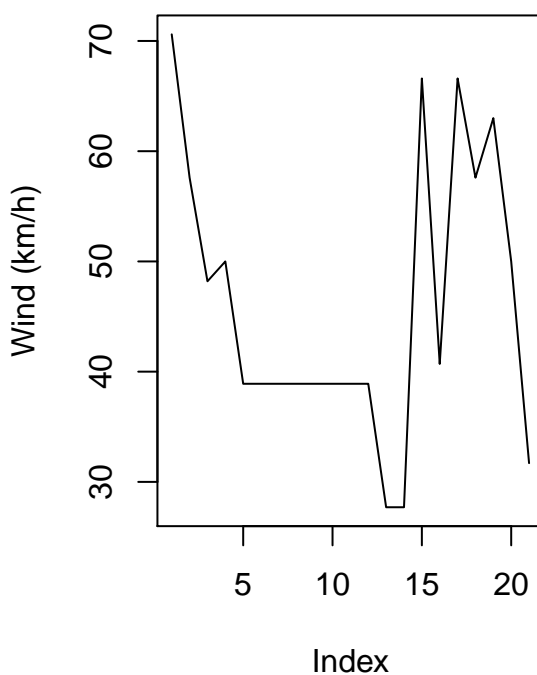
# Check imputation
par(mfrow = c(1, 2))
plot(ts_df[140:160, "wind"], type = "l", ylab = "Wind (km/h)", main = "Raw Data (Index
↪ 140 to 160)")
plot(ts_imputed[140:160, "wind"], type = "l", ylab = "Wind (km/h)", main = "After Linear
↪ Imputation")

```

Raw Data (Index 140 to 160)

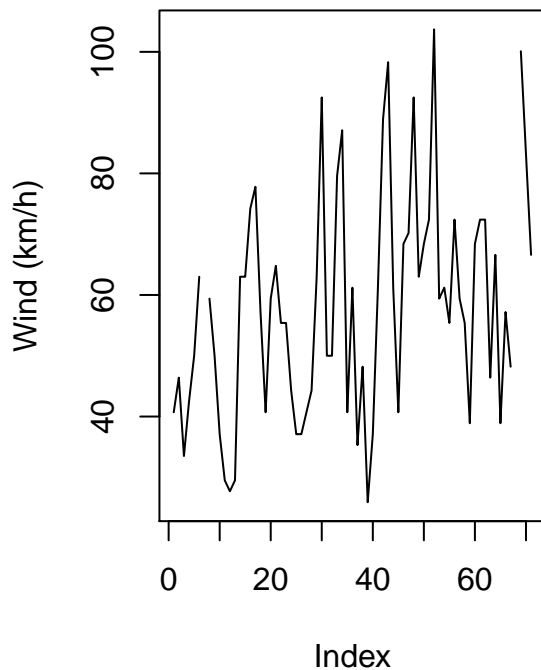


After Linear Imputation

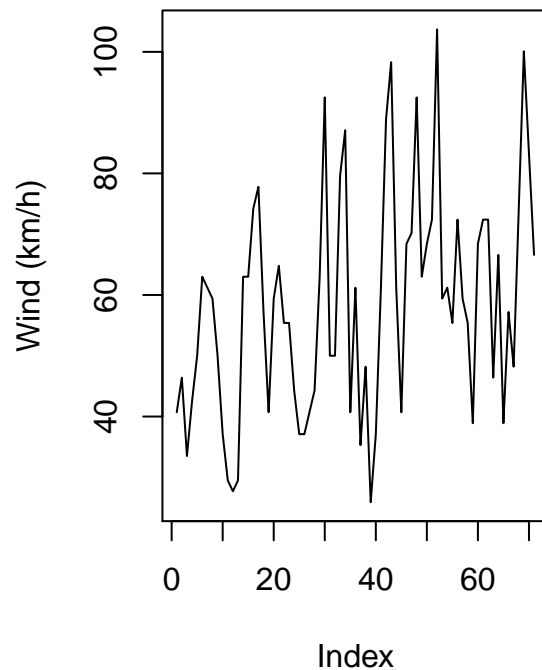


```
par(mfrow = c(1, 2))
plot(ts_df[3610:3680, "wind"], type = "l", ylab = "Wind (km/h)", main = "Raw Data (Index
↪ 3610 to 3680)")
plot(ts_imputed[3610:3680, "wind"], type = "l", ylab = "Wind (km/h)", main = "After
↪ Linear Imputation")
```

Raw Data (Index 3610 to 3680)



After Linear Imputation



Train and Test Set Splitting

```
# split train and test set
split_index = length(ts_imputed[, 6]) * 0.8

lower = ts_imputed[, 6][round(split_index)]
upper = ts_imputed[, 6][round(split_index) + 1]

train = ts_imputed[ts_imputed[, 6] <= lower, ]
test = ts_imputed[ts_imputed[, 6] >= upper, ]
```

```
# Check test and train set ratio
length(test)/length(ts_imputed) # 20%
length(train)/length(ts_imputed) # 80%
length(train) + length(test) == length(ts_imputed)

# Check start and end dates of test data
as.Date(train[, "date"])[6034]
```

```
train_df = train %>% as.data.frame() %>% mutate(date = as.Date(date))
test_df = test %>% as.data.frame() %>% mutate(date = as.Date(date))
```

```
write.csv(train_df, "train.csv", row.names = FALSE)
write.csv(test_df, "test.csv", row.names = FALSE)
```